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Enhancing Image Deblurring with Advanced Learning

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Abstract: This study presents a novel framework for adaptive photo restoration by integrating GFPGAN for face-focused enhancement and Real-ESRGAN for general image refinement. Users select modes tailored to image content. The model features blur-aware preprocessing, intelligent background boosting, and output evaluation through SSIM. The application is deployed using a user-friendly Gradio interface and shows consistent performance across varied visuals.

I. INTRODUCTION

Recent innovations in neural networks have revolutionized the way degraded visuals are enhanced. Traditional models operate on single use cases. Our architecture bridges this gap by intelligently routing images through models that specialize in either facial clarity or broad scene enhancement.

II. LITERARURE SURVEY

Wang et al. (2021) implemented Real-ESRGAN to reconstruct fine textures in natural scenes. GFPGAN tackled real-world facial degradation challenges.

Lim et al. (2017) proposed a residual learning approach for resolution boosting. Isola et al. explored translation models for visual domain shifts.

III. EXISTING SYSTEM

Current approaches apply a fixed enhancement model, often failing when input complexity varies. Tools lack preprocessing intelligence, model fallback, and don't adapt enhancement logic.

IV. PROPOSED SYSTEM

We developed a two-mode enhancement app. Users select the image type. The system applies pre-enhancement filtering based on clarity. Facial images use GFPGAN, while Real-ESRGAN handles landscapes. Mask-based enhancement isolates regions to avoid distortion.

V. IMPLEMENTATION

Using Python and libraries like OpenCV, NumPy, and Gradio, the models were integrated into an interface that guides image flow from input to result visualization. The models include pretrained GFPGAN v1.3 and Real-ESRGAN Anime x4.

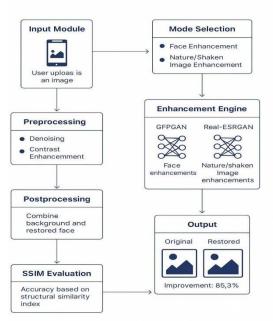
VI. MODULES

Image Uploader \rightarrow Mode Selector \rightarrow Smart Preprocessor \rightarrow Enhancement Core (GFPGAN/ESRGAN) \rightarrow Optional Masking \rightarrow Result Generator \rightarrow SSIM Evaluator.

VII. ALGORITHMS

Custom logic calculates blur via Laplacian variance. LAB space is used for contrast tuning. GAN models generate refined outputs. SSIM provides restoration feedback.





VIII. SYSTEM ARCHITECTURE

IX. RESULTANDANALYSIS

Ultimate Photo Enhancer

Upload a photo and choose a mode! Face Enhancement uses GFPGAN for faces, Nature/Shaken Image Enhancement uses ESRGAN for landscapes and shaken pics.

65. Upload a Photo	53 Original vs Restored Photo		
Drop Image Here			
<u>.</u> e C	Improvement Estimate		
Enhancement Mode			
Face Enhancement •	Share via Link		
Clear Submit			

Fig1. Opening Interface

ganize - New folder			-	0
Tejeswar - Perso Attachments Documents Pictures 00		I		
Desktop Downloads Documents 2 Documents 2	a a		Select a file to preview.	
File name:		~	Image Files Open Cancel	~
	<u>e</u> © 1			
Enhancement Mode				
Face Enhancement			÷.	
Clear			Submit	

Fig2.openingafile



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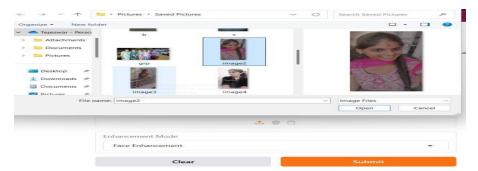


Fig3.selectingan image

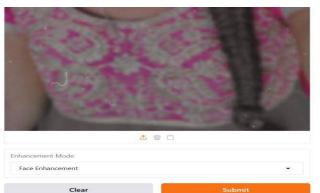


Fig4.selectingenhancementmode



Fig5.OriginalimagevsRestoredimage Fig6.OriginalimagevsRestoredimagefor Nature or Shaken imagefor Face Enhancement Visual results show marked improvement in detail clarity. Restoration improvements were evaluated using SSIM and user feedback.

X. CONCLUSION

A modular photo enhancement system was created by combining two specialized Generative AI models. This platform achieves accurate and high-quality restoration for various image types.

XI. FUTURE SCOPE

Automation in mode selection, mobile deployment, and multilingual UI support are planned for broader accessibility.

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